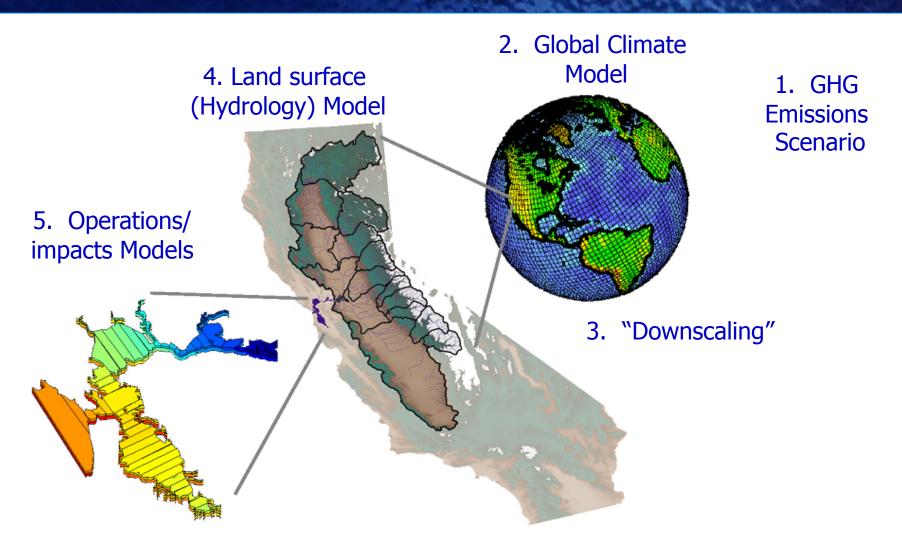


IGIM CMIP meeting College Park, MD October 4, 2012

# A well-meaning hydrologist walks into a climate change study...

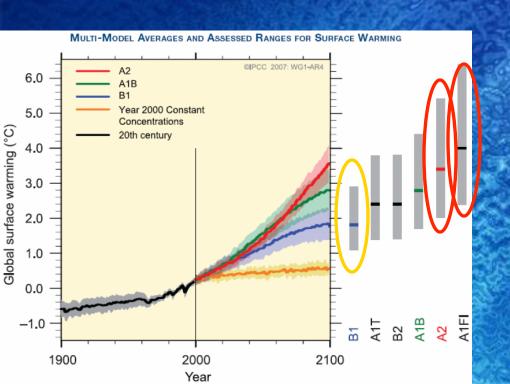


# Estimating Climate Impacts to Water



## Selecting GCM runs: "Bookends"

- Brackets range of uncertainty
- Useful where impacts models are complex
- Downscale output from few GCMs



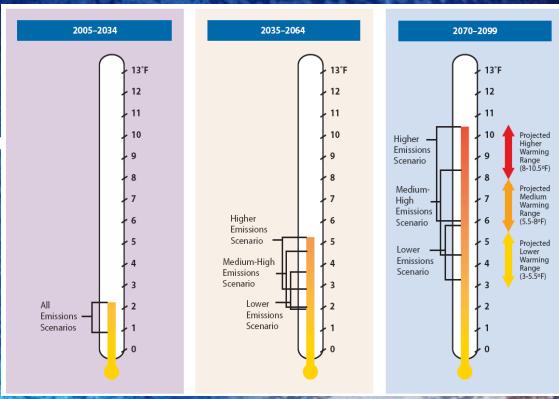
12 1 10 10 1	
AOGCM	Equilibrium climate sensitivity (°C)
1: BCC-CM1	n.a.
2: BCCR-BCM2.0	n.a.
3: CCSM3	2.7
4: CGCM3.1 (T47)	3.4
5: CGCM3.1 (T63)	3.4
6: CNRM-CM3	n.a.
7: CSIRO-MK3.0	3.1
6: ECHAM5/MPI-OM	3.4
9: ECHO-G	3.2
10: FGOALS-g1.0	2.3
11: GFDL-CM2.0	2.9
12: GFDL-CM2.1	3.4
13: GISS-AOM	n.a.
14: GISS-EH	2.7
15: GISS-ER	2.7
16: INM-CM3.0	2.1
17: IPSL-CM4	4.4
18: MIROC3.2(hires)	4.3
19: MIROC3.2(medres)	4.0
20: MRI-CGCM2.3.2	3.2
21: PCM	2.1
22: UKMO-HadCM3	3.3
23: UKMO-HadGEM1	4.4

#### **Bookend results for California**

- CA average annual temperatures for 3 30-year periods
- Amount of warming depends on our GHG emissions at end of 21<sup>st</sup> century.
- Summer temperatures increases (end of 21st century) vary widely:

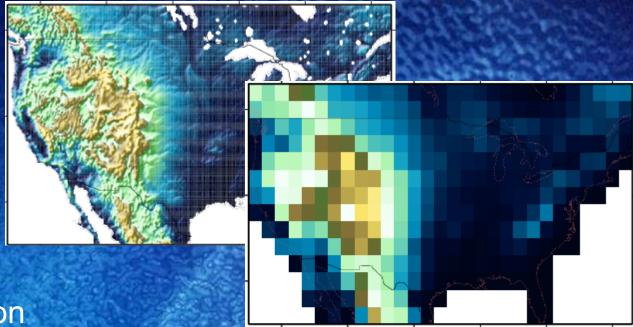
Lower: 3.5-6 °F

Higher: 6-10.5 °F

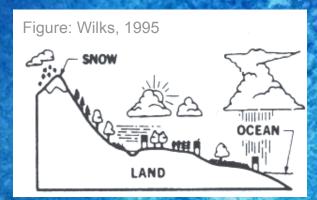


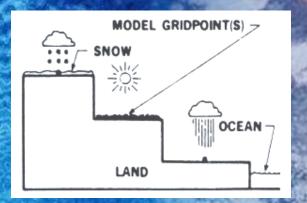
# Downscaling: bringing global signals to regional scale

GCM scale and processes at too coarse a scale



- Resolved by:
  - -Bias Correction
  - -Spatial Downscaling

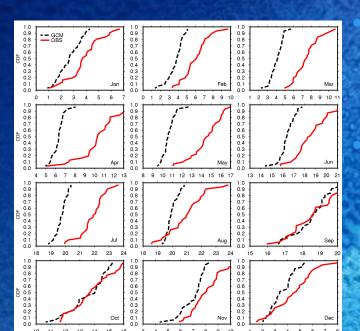




### **BCSD Method – "BC"**



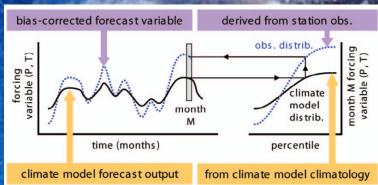
- At each grid cell for "training" period, develop monthly CDFs of P, T for
  - GCM
  - Observations (aggregated to GCM scale)
  - Obs are from Maurer et al. [2002]



Use quantile mapping to ensure monthly statistics (at GCM scale) match obs

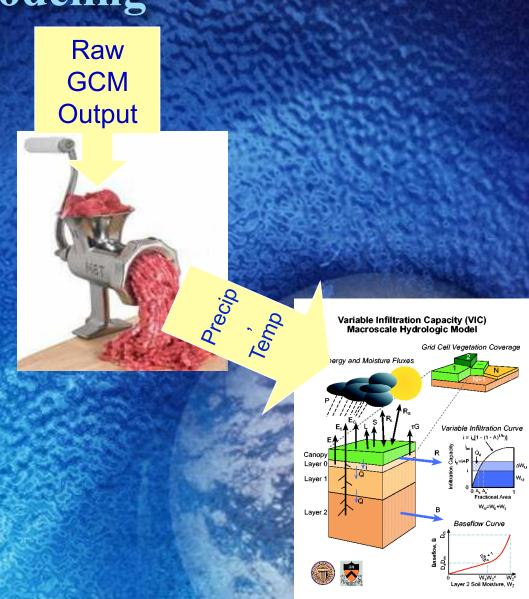
Apply same quantile mapping to "projected" period

Wood et al., BAMS 2006



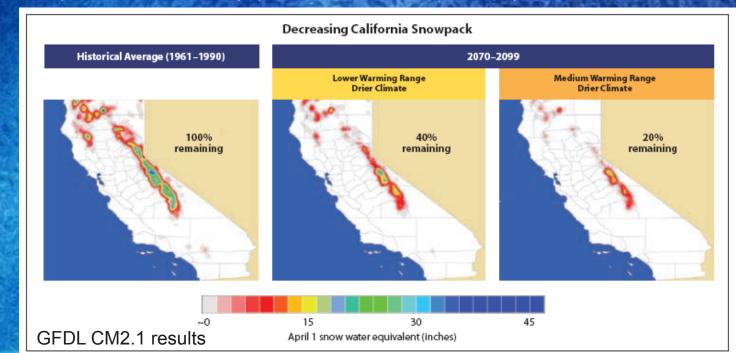
# Downscaling for Hydrology Impact Modeling

- BCSD downscaling of GCM Precip and Temp
- Use to drive VIC model
- Obtain runoff, streamflow, snow



## **Projected Impacts: Loss of Snow**

- Snow water in reserve on April 1
- Change (Sacramento-San Joaquin basin, 2 GCMs, 2 emissions scenarios):
  - -12% to -42% (for 2035–2064) (up to 1 Lake Shasta)
  - -32% to -79% (for 2070–2099) (up to 2 Lake Shastas)



### Some Agency and Organizational Responses

World Federation of Engineering Organizations (2009)	To develop and implement engineering tools, policies and practices for risk assessment and adaptation of existing and new civil infrastructure to climate change
Water Utility Climate Alliance (CAP, Denver Water, MWD, NYC DEP, SFPUC and others)	Collaborating on climate change issues affecting drinking water utilities.
Federal Climate Change and Water Working Group (Reclamation, USACE, NOAA and USGS) (2008)	Helping the water management community adapt practices as climate changes
ASCE - Committee on Adaptation to a Changing Climate (2010)	Encourage assessments of the built and natural environment to find and quantify vulnerabilities and incorporate updated criteria into engineering practice
California Climate Action Team - Water-Energy Team (2005)	Coordinating GHG emission reduction and adaptation actions affecting energy that supports the storage, transport and delivery of water
Santa Clara Valley Water District	Climate change addressed in watershed and flood control planning

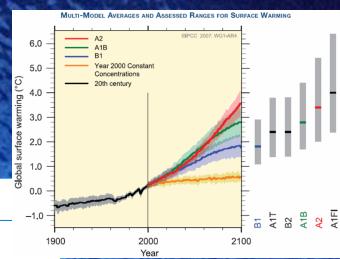
Background: Confederation Bridge in the Gulf of Saint Lawrence (http://www.cakex.org)

### **IPCC CMIP3 GCM Simulations**

- 20<sup>th</sup> century through 2100 and beyond
- >20 GCMs
- Multiple Future Emissions Scenarios

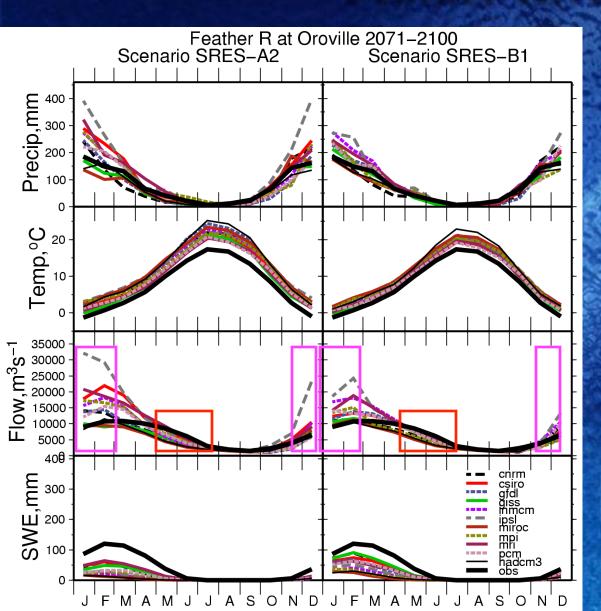


	Plentrl	PDcntrl	20C3M	Commit	SRESA2	SRESA1B	SRESB1	1%to2x	1%to4x	Slabontl	2xCO2	AMIP
BCC-CM1, China		2	4				2	1	1			4
BCCR-BCM2.0, Norway	1		11	1	1	1	1_	1				
CCSM3, USA	2	1	9	5	5	7	œ O	1	1	1	1	1
CGCM3.1(T47), Canada	1		5	5	5	5	4	1	1	1	1	
CGCM3.1(T63), Canada	1		1			1	1	1		1	1	
CNRM-CM3, France	1	(		1 (		1	$\left( \right)$	1	1			1
CSIRO-Mk3.0, Australia	2		ß	1		1		1		1	1	
CSIRO-Mk3.5, Australia	1		1	1	1	1	)-(	1				
ECHAM5/MPI-OM, Germany	1		$\bigcirc \blacktriangle$	3	3	4	)	3	1	1	1	3
ECHO-G, Germany/Korea	1	1	5	4	3	3	${2}$	1	1			
FGOALS-g1.0, China	3		3	3		3	3	3				3
GFDL-CM2.0, USA	1	(	3	1		1	$\overline{\mathcal{I}}$	1	1	1	1	
GFDL-CM2.1, USA	1		3	1	1	1	1	1	1			
GISS-AOM, USA	2	(	$\sim$ 2			2	$\bigcirc$ 2					
GISS-EH, USA	1		5			4		1				
GISS-ER, USA	1		9	1	1	5	1	1	1	1	1	4
INGV-SXG, Italy	1		1		1	1		1	1			
INM-CM3.0, Russia	1	(	$\bigcup$	1		1		1	1	1	1	1
IPSL-CM4, France	1	1	2	1		1		1	1			6
MIROC3.2(hires), Japan	1		1			1	1	1		1	1	1
MIROC3.2(medres), Japan	1		3	1	3	3	3	3	3	1	1	3
MRI-CGCM2.3.2, Japan	1	1	5	1	5	5	5	1	1	1	1	1
PCM, USA	1	1	4	3	4	4	4	5	1			1
UKMO-HadCM3, UK	2	(	2	1		1		1				
UKMO-HadGEM1, UK	1		1		1	1		2	1	1	1	1



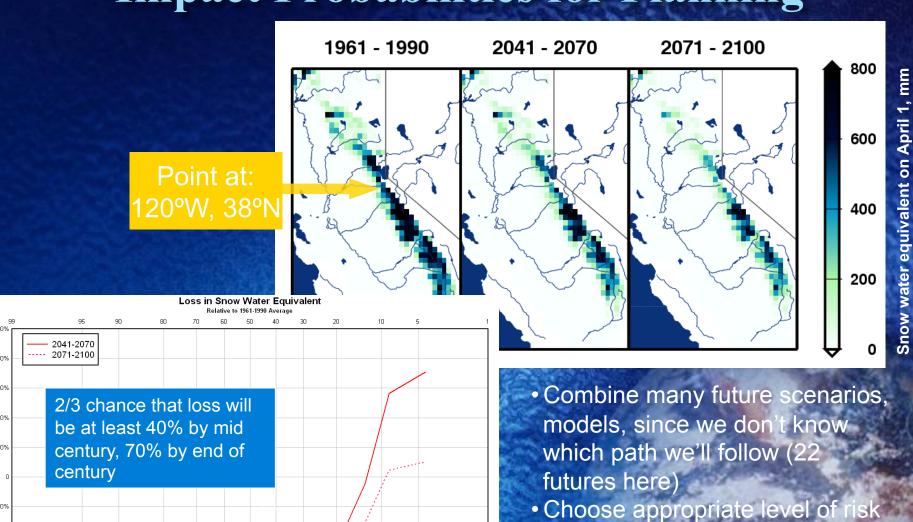


# Multi-Model Ensemble Projections for Feather River



- Increase Dec-Feb Flows+77% for A2+55% for B1
- •Decrease May-Jul -30% for A2 -21% for B1

## Impact Probabilities for Planning



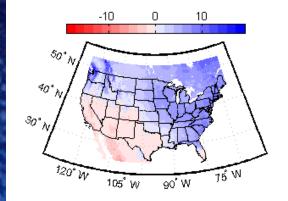
## Demand for downscaled data

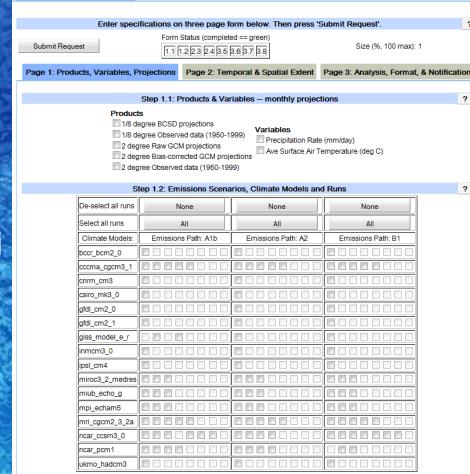


## Monthly downscaled data

- PCMDI CMIP3 archive of global projections
- 16 GCMs, 3 Emissions
- 112 GCM runs
- Allows quick analysis of multi-model ensembles
- gdo4.ucllnl.org/ downscaled\_cmip3\_projections





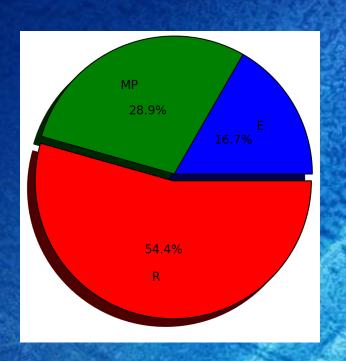


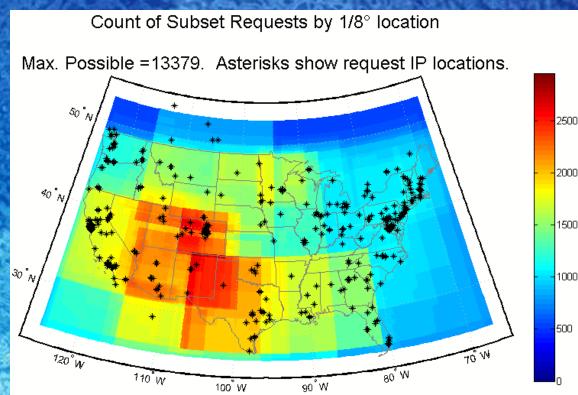
BCCA-CMIP3-Climate-daily

BCSD-CMIP3-Climate-monthly

### Use of U.S. Data Archive

- Thousands of users downloaded >20 TB of data
- Uses for Research (R), Management & Planning (MP), Education (E)





# What is missing from downscaled data archive?

Items Requested		
Daily Data		
Additional Climate Variables:		
Tmax/Tmin		
Evapotranspitation/Potential Evapotranspiration		
Surface Wind		
Humidity		
Solar Radiation		
Surface Pressure		
Cloud Cover		
Derived Variables:		
Runoff		
Snow		
Soil Moisture		
Raw and intermediate data (from the downscaling process)		
More resources to cope with netCDF files		
Data from other downscaling methods to intercompare		
Full historic data for 20th century (1900-current)		
Ability to extract watersheds (not just rectangular subsets)		
Expanded spatial domain		

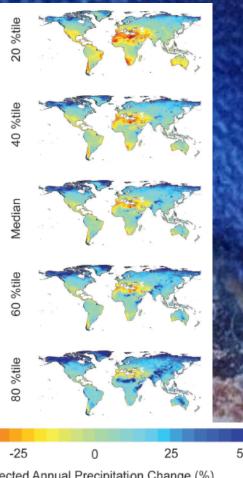
### Global BCSD

- Similar to US archive, but ½-degree
- Publicly available since 2009
- Captures variability among GCMs
- www.engr.scu.edu/~emaurer/global\_data/
- Data accessed by users in all 50 States and 99 countries (last 11 months only)



Visits from 3 Nov 2011 to 27 Sep 2012





Projected Annual Precipitation Change (%) 2070-2099 Compared to 1961-1990 Baseline

Source: Girvetz et al, PloS, 2009

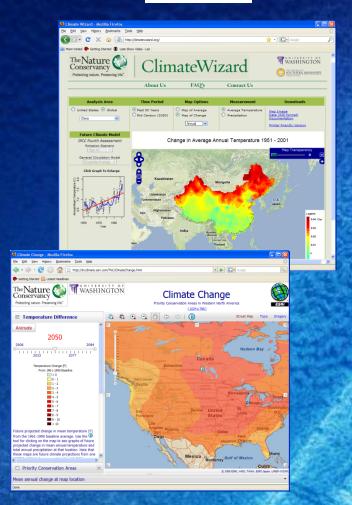
-50

# Most commonly requested items

Items Requested			
Daily Data			
Additional Climate Variables:			
Tmax/Tmin			
Evapotranspitation/Potential Evapotranspiration			
Surface Wind			
Humidity			
Solar Radiation			
Surface Pressure			
Cloud Cover			
Derived Variables:			
Runoff			
Snow			
Soil Moisture			
Raw and intermediate data (from the downscaling process)			
More resources to cope with netCDF files			
Data from other downscaling methods to intercompare			
Full historic data for 20th century (1900-current)			
Ability to extract watersheds (not just rectangular subsets)			
Expanded spatial domain			

# Online Analysis and Download with http://ClimateWizard.org

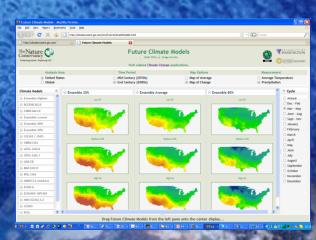
#### Level 1



#### Level 2



#### Level 3



- Global and US data sets
- Country and US state
- boundaries defined
- Spatial and time series
- analysis
- Upload of custom shapefile

Girvetz et al., PLoS, 2009

### **Too much information?**

Little guidance in selection of: Emissions GCMs

Hundreds of downscaled GCM runs

Many impacts studies cannot use all of them

How much information is really useful?



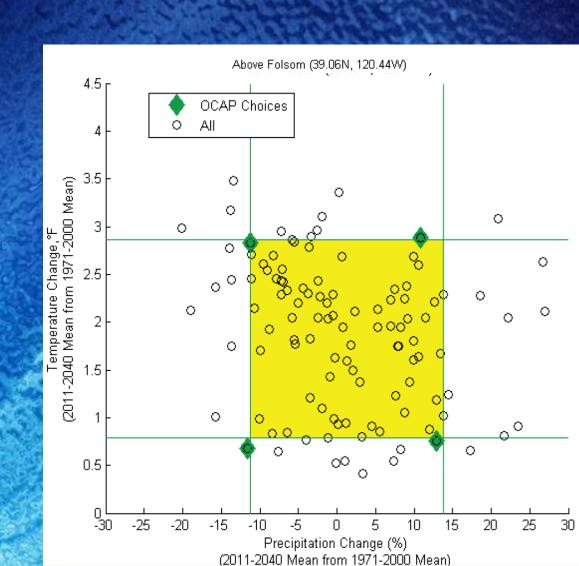
### Selecting Specific GCM Runs

Bivariate probability plot shows correlation between  $\Delta T$ ,  $\Delta P$ 

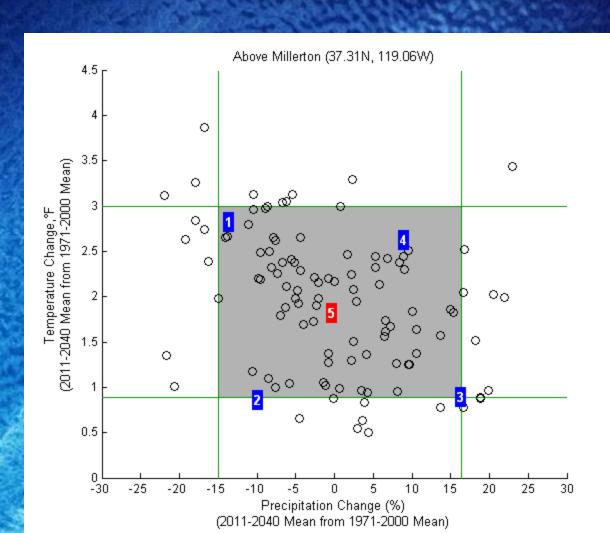
Identify Change Range: 10 to 90 %-tile ΔT, Δ P

Select bounds based on:

- risk attitude
- interest in breadth of changes
- number of simulations desired

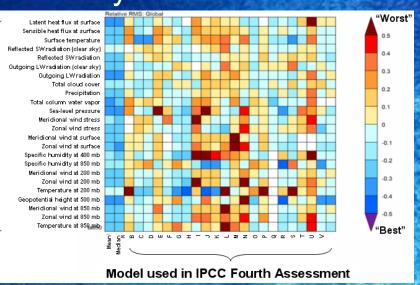


### Or use 5?

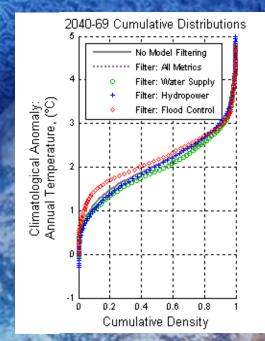


## Selecting GCMs for Impact Studies

- Ensemble mean provides better skill
- Little advantage to weighting GCMs according to skill
- Most important to have "ensembles of runs with enough realizations to reduce the effects of natural internal climate variability" [Pierce et al., 2009]
- Maybe 10-14 GCMs is enough?

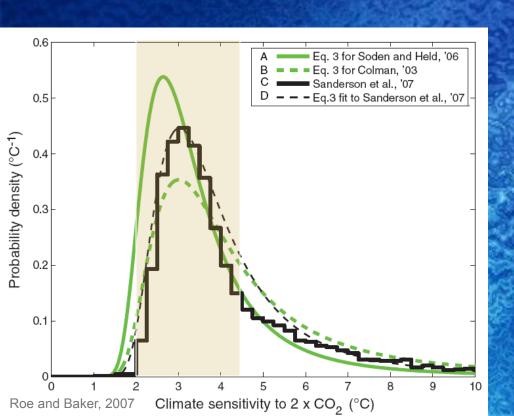


Climate variable



# Do CMIP GCM runs capture important uncertainties?

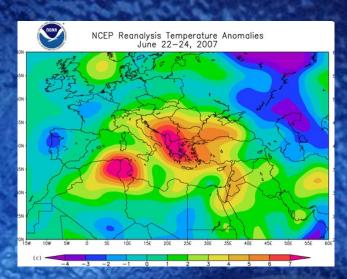
- Perturbed physics ensembles
- Is planning for the higher probability outcomes appropriate?



AOGCM	Equilibrium climate sensitivity (°C)
1: BCC-CM1	n.a.
2: BCCR-BCM2.0	n.a.
3: CCSM3	2.7
4: CGCM3.1(T47)	3.4
5: CGCM3.1(T63)	3.4
6: CNRM-CM3	n.a.
7: CSIRO-MK3.0	3.1
8: ECHAM5/MPI-OM	3.4
9: ECHO-G	3.2
10: FGOALS-g1.0	2.3
11: GFDL-CM2.0	2.9
12: GFDL-CM2.1	3.4
13: GISS-AOM	n.a.
14: GISS-EH	2.7
15: GISS-ER	2.7
16: INM-CM3.0	2.1
17: IPSL-CM4	4.4
18: MIROC3.2(hires)	4.3
19: MIROC3.2(medres)	4.0
20: MRI-CGCM2.3.2	3.2
21: PCM	2.1
22: UKMO-HadCM3	3.3
23: UKMO-HadGEM1	4.4

## **Downscaling for Extreme Events**

- Some impacts due to changes at short time scales
  - Heat waves
  - Flood events
- Daily GCM output limited for CMIP3, more plentiful for CMIP5
- Downscaling adapted for modeling extremes



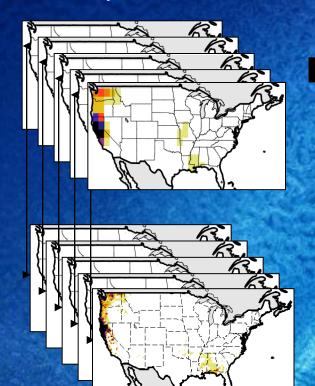


# Most commonly requested items

Items Requested		
Daily Data		
Additional Climate Variables:		
Tmax/Tmin		
Evapotranspitation/Potential Evapotranspiration		
Surface Wind		
Humidity		
Solar Radiation		
Surface Pressure		
Cloud Cover		
Derived Variables:		
Runoff		
Snow		
Soil Moisture		
Raw and intermediate data (from the downscaling process)		
More resources to cope with netCDF files		
Data from other downscaling methods to intercompare		
Full historic data for 20th century (1900-current)		
Ability to extract watersheds (not just rectangular subsets)		
Expanded spatial domain		
Expanded spatial domain		

## **Constructed Analogues**

Library of previously observed anomaly patterns:





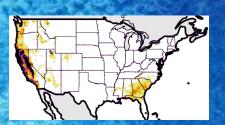
Coarse resolution analogue:



Analogue is linear combination of best 30 observed

Given daily

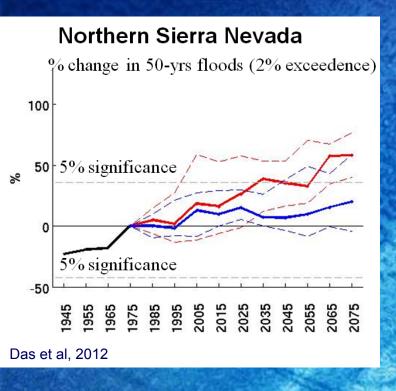
**GCM** anomaly

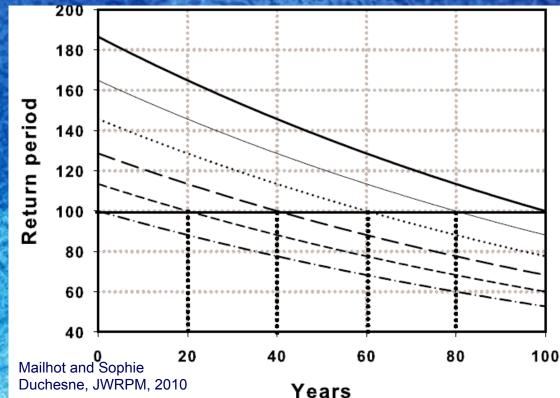


Apply analogue to fine-resolution climatology

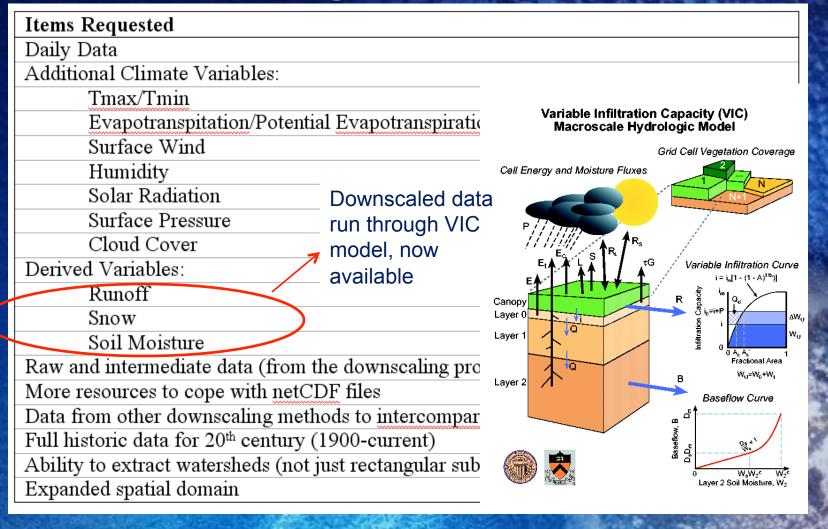
# Sustainable Design in a Dynamic Environment

- Declining return periods for extreme events
- A solution: Overdesign for present





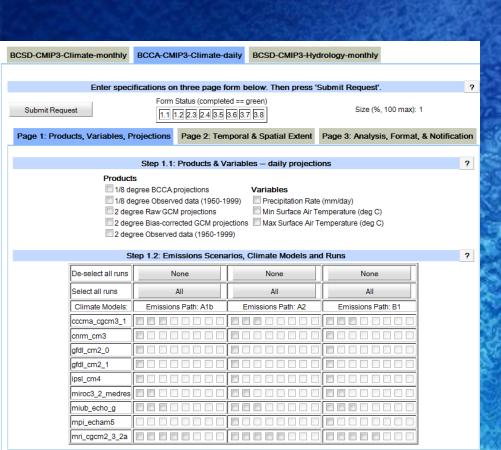
# What is missing from downscaled data original archive?

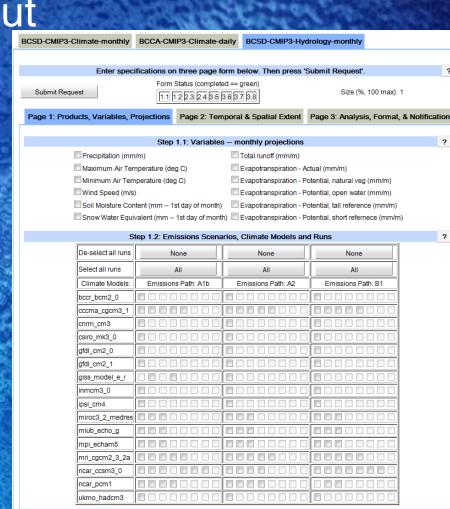


## Archive expansion (still CMIP3)

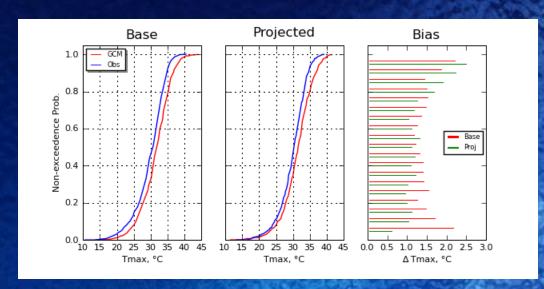
Daily downscaled data

Hydrology model output



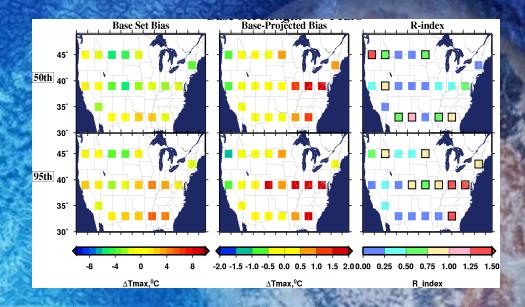


### Is bias correction effective?



Biases vary in time, space, at quantiles

- On average, bias correction works
- But for small ensembles maybe not

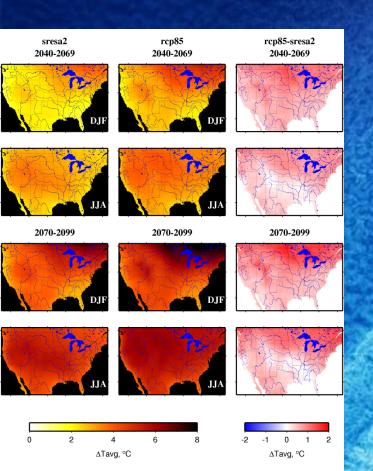


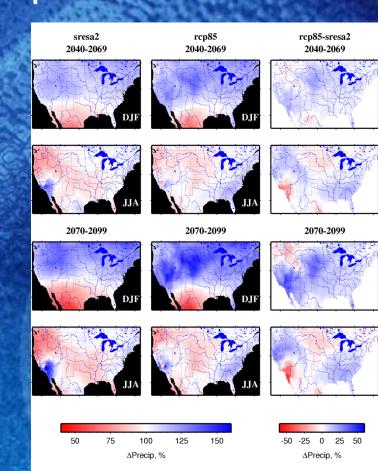
### CMIP5 additions to archive

- Monthly downscaling of Tmax, Tmin, Precip for:
  - 84 historical GCM runs
  - 237 projections (total for 4 RCPs)
- Daily downscaling with two techniques:
  - 46 historical runs
  - 147 projections (total for 4 RCPs)
- Hydrology model output for 100 runs

# Does CMIP3 or CMIP5 choice matter?

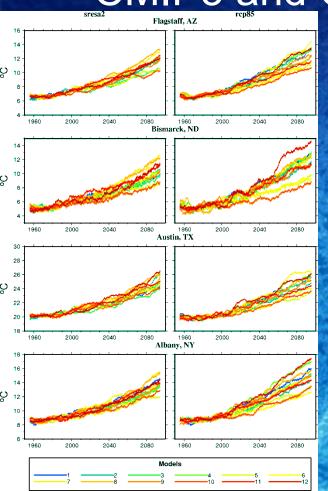
- Ensemble average changes comparable
- RCP8.5 and SRES A2 comparable

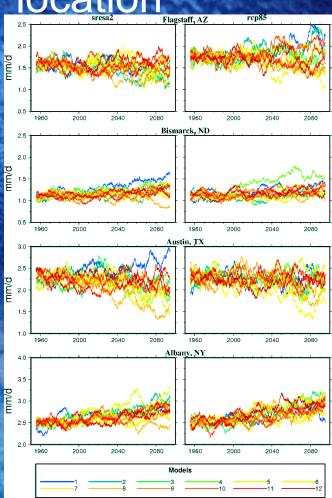




## **Model Spread**

 Differences in model spread between CMIP3 and CMIP5 varies by location





### Information overload overload

- If 112 GCM projections wasn't too much, is 500?
- Have we progressed in providing policymakers with information for...
  - Selecting concentration pathways
  - Assembling an ensemble of GCMs
  - Using appropriate downscaling
  - Interpreting results
- Can we (conditionally) recommend anything?

